

Reconstructing the Magnetic field of the Milky way via Astrophysical Techniques and Numerical Simulations (MAGMASIM)

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Outline of the talk

- Introduction
- Reconstruction / Calculation strategy
- Initial Results
- Conclusions and Perspectives

Introduction: Scope of the project

• Aim of the Project

Reconstructing the magnetic field of our Galaxy (GMF)

Synergy between the Institute of Applied and Computational Mathematics (IACM) and

the Institute of Astrophysics (IA) of FORTH

What is the "reconstruction"?

Using

> constraints from experimental / observational measurements of the Galactic Magnetic Field and

modern numerical tools

calculate the GMF in the all points in a geometry of interest (cone)

Introduction: Galactic Magnetic Field

Why the study of GMF is important?

- Affects Star Formation
- ✤ Affects the *trajectory of high-energy cosmic rays* ~10²⁰eV (charged high-energy particles) that give us information about high-energy astrophysical phenomena
- May give us information for the *first moments of the universe*

Basic Theoretical Tool for Magnetic Field Calculations

- Maxwell Differential Equations with Boundary Conditions (Values of the MF in the boundaries of the geometry of the calculation)
- They reduce to the Poisson equation (MF is time independent no sources)

Introduction: Our Galaxy (Milky Way)



The structure of the Milky Way is thought to be similar to this galaxy (image by Hubble)

1 Light Year = $9.46 \times 10^{15}m$ 1 parsec= $3.086 \times 10^{16}m$

- Spiral Galaxy
- Radius: 130000 light years
- Total Mass: 1.5 trillion Solar Masses
- Contains between 100 and 400 billion stars and at least that many planets
- The Galactic disc consists of a stellar component (composed of most of the galaxy's stars) and a gaseous component (mostly composed of cool gas and dust)
- The Galactic disk is surrounded by a *spheroidal halo* of old stars
- In the Galactic Center there is central object thought to be a supermassive Black Hole
- The *Sun* is ~8.5kpc away from the Galactic Center
- Permeated by Magnetic Field

Introduction: Galactic Magnetic Field

Magnetic fields are prevalent in the Milky Way





Image by NASA MFs are mostly Galactic (stars, planets relatively small scales) (i.e. Earth's $\sim 0.5G$) Image by Planck Telescope The GMF is of larger scale *Observations*: $\sim 6\mu G$ near the Sun, $\sim 30\mu G$ as we approach the Galactic center

Calculation Strategy: GMF Reconstruction

Numerical Simulation (solution of the Maxwell Equations) subject to constrains from experimental measurements

- Till now: Integral measurements along the line of sight (not trivial to include in calculations)
- Direct measurements of the GMF are only now starting to become available and an influx of them is expected in 5 to 10 years

State of the art: Assume a geometry for the GMF and perform a best-fit that reproduces the integral measurements

Such models are too simple to capture the full complexity of the GMF

GMF Reconstruction - Challenges

Numerical challenges:

- Observational measurements will be used as boundary conditions (BC) or sparse known data points in the computational grid
- However, direct measurements of the GMF are only possible at the locations of interstellar clouds —> no experimental control over them. The GMF will be known on an irregular and very sparse grid
- Integral constrains from the *line-of-sight* measurements must be included in the model

GMF Reconstruction – Tackling the problem

To overcome the previous challenges, our approach involves various stages

- *The direct (forward) problem*: We solve numerically the differential equations, describing the problem, in a 3D geometry given specific boundary conditions (BCs)
- The Sparse Data case Inverse problem: From the data created in the previous step we generate simulated direct observational data, and we try to infer the BCs by solving the inverse problem
- Inclusion of *Integral Constrains* that simulate the along the line-of-sight measurements
- Application to real *direct observational data*

GMF Reconstruction – Methodology – Results

First Stage: **"The direct – forward problem"**

- A cone geometry is used to model the realistic geometry of the experimental observations
- Solve the differential equations with modern tools (finite elements)
 We use FEniCS
- A popular open-source computing platform for solving partial differential equations
- Enables users to quickly translate scientific models into efficient finite element code in Python

GMF Reconstruction – Results

First Stage: **"The direct – forward problem"**

An example calculation

- Poisson equation: $\nabla^2 \mathbf{B} = 0, \mathbf{x} \in \text{in cone}$
- Analytical boundary conditions $B_x = 1 + x^2 - 2y^2 + z^2$

(similar functions for the other components)

• *Reference calculation*: we will use this calculation it to **simulate artificial experimental data**



An example plot of the x-component of the Magnetic Field. Other components produce similar results

GMF Reconstruction – Results

Second Step: "The inverse problem"

- Consider a set sparse direct measurements of the MF (created from the solution of the direct problem or, in the future, from actual observational data)
- From them we want to find information for the statistical properties of the boundary conditions

This defines the inverse problem

- As an example, consider the *simulated* data calculated by solving the direct problem using the shown boundary conditions
- The BC are created by two normal distributions one for the upper and one for the lower half of the cone with different mean values



GMF Reconstruction – Results – A two prior example

Second Step: "The inverse problem"

To infer information for the boundary conditions we use **Bayesian Analysis**

 $p(\theta|y) \propto p(y|\theta)p(\theta)$

 $p(\theta)$: Prior(s), 2 in this example, distributions that describe the statistical properties of the BC (here we consider normal distributions).

The **unknown** parameters $\boldsymbol{\theta}$ can be mean values of the distributions etc.

Using values generated from the priors as BC we solve the direct problem with *FEniCS*

 $p(y|\theta)$: Likelihood, usually a normal distribution (y is the difference between this and the simulated data set) $p(\theta|y)$: Posterior



GMF Reconstruction – Results – A two prior example

Second Step: "The inverse problem"

To find *θ* we define and solve (with proper optimization numerical methods like simulated annealing) a *Maximum Likelihood Estimation* problem:

$$\theta_{\text{MLE}} = \arg \max_{\theta} (p(\theta \mid y))$$

In this example, with the help of 2 prior distributions, the statistical properties of the initial distributions are successfully recovered



GMF Reconstruction – Current Work

Complete the study of the inverse problem

- Study the case of more priors
- See how the sparsity of the data affects the results

Incorporate the information from the line-of-sight measurements. This is done with the introduction of integral constrains in the calculation

Future: Apply the inverse problem algorithm with actual observational data once available

Conclusions and Perspectives

✓ The problem of the numerical reconstruction of the GMF is very interesting from both the Astrophysical and the Numerical point of view

✓ The reconstruction will be very useful for the study of all astrophysical phenomena affected by Galactic magnetism

The problem, from the numerical point of view, is very challenging but also very interesting mainly due to sparse data and integral constrains Thank you for you attention